





Detecting and tracking marine mammals around tidal turbines: development of a dual multibeam sonar system

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Thanks...





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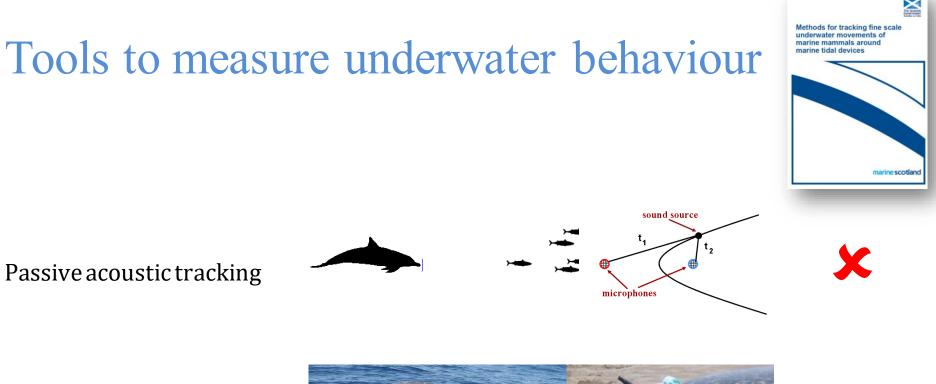


Energetic habitats: important for tidal turbines and seals





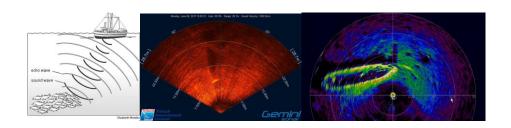




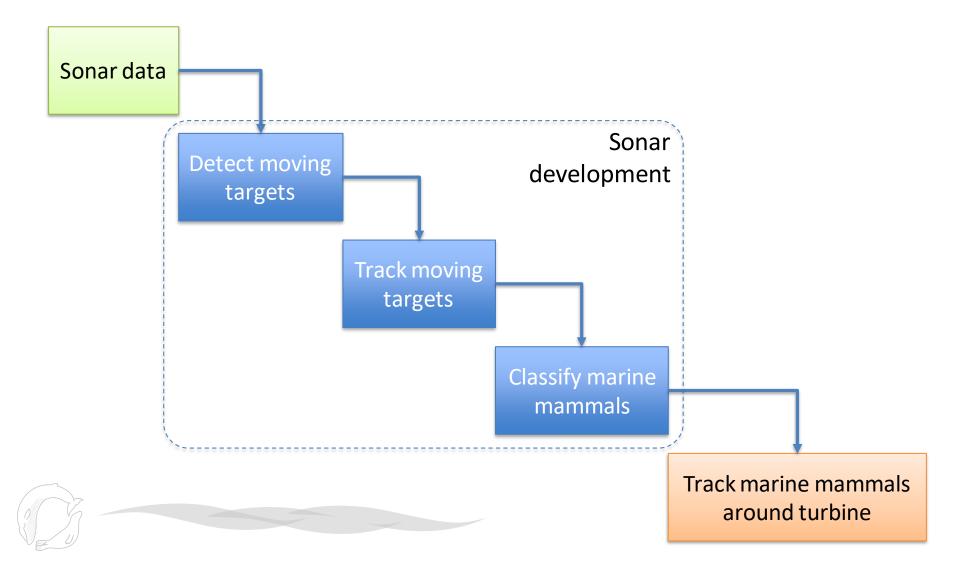
Animal borne tags/telemetry







Tracking marine mammals with sonar: the process



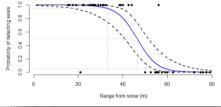
Sonar data



Tritech Gemini: 720 kHz multibeam

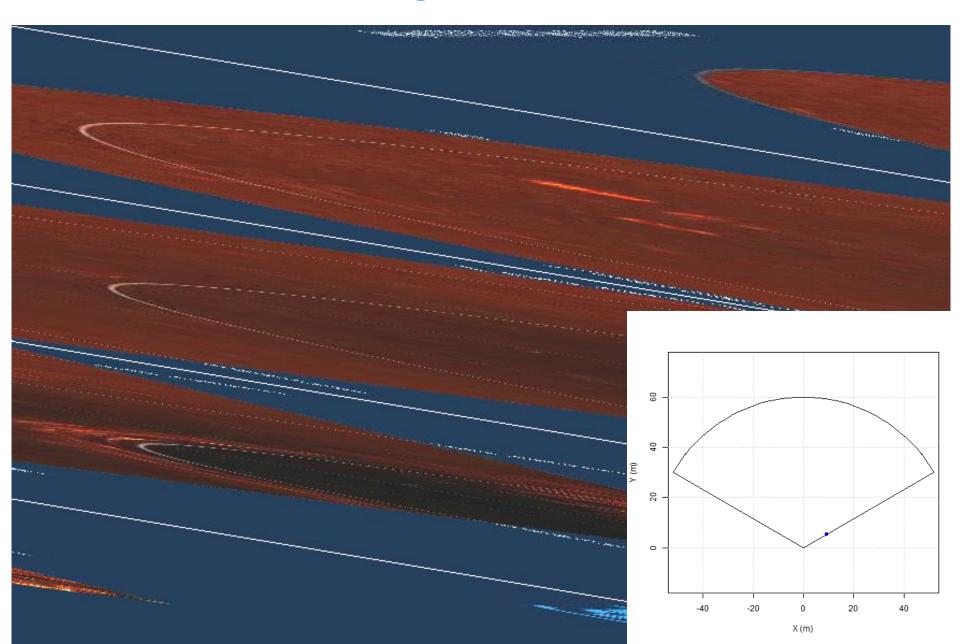
No measured responses to signals by seals

Existing target tracking module



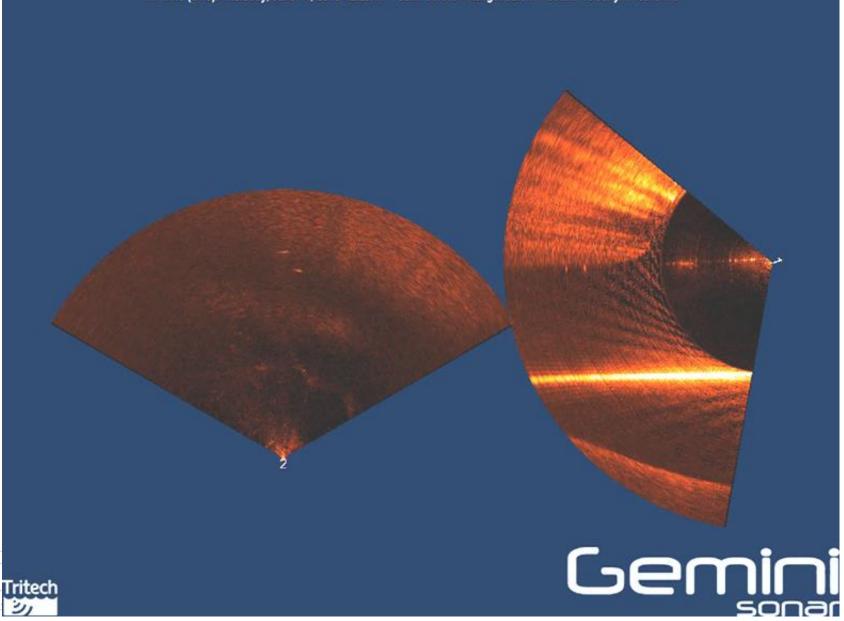
Detection probability (seals) >0.95 up to 30m

Detect and track targets



Detect and track targets – 3D

Tx Time (UTC): Thursday, June 11, 2015 12:23:47 Gain: 80.0% Range: 55.0m Sound Velocity: 1489.6m/s



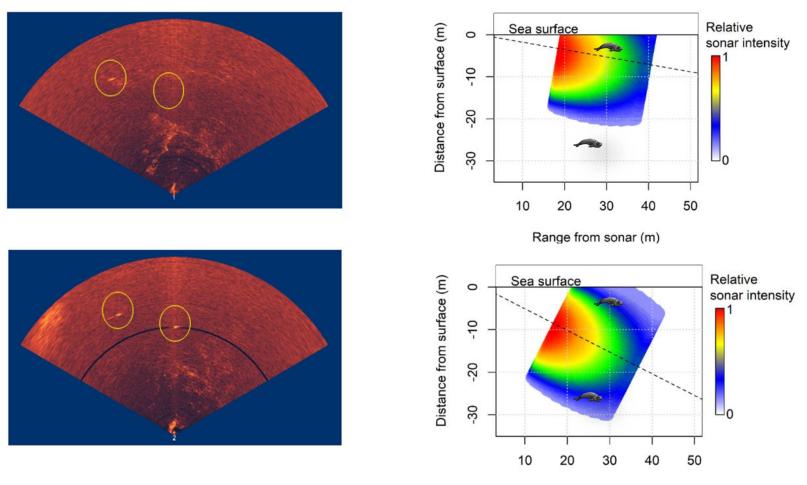
Detect and track targets – 3D

Tx Time (UTC): Wednesday, June 10, 2015 19:57:11 Gain: 100.0% Range: 46.0m Sound Velocity: 1489.6m/s





Detect and track targets – 3D

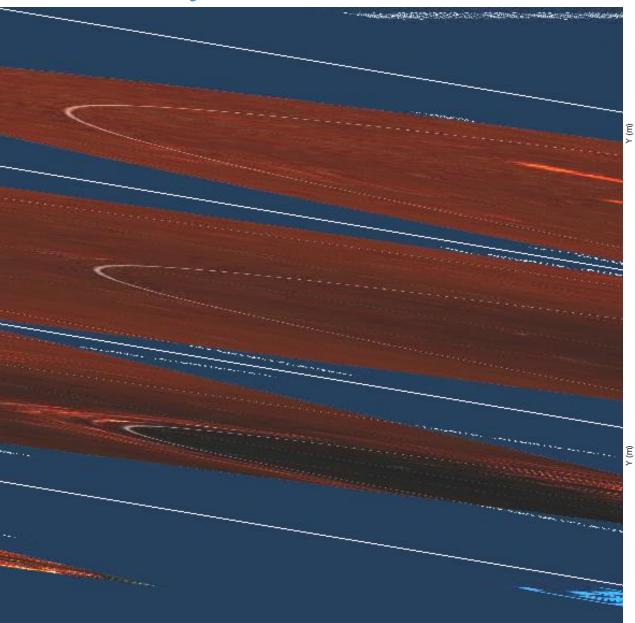


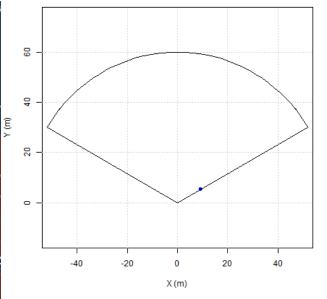
Range from sonar (m)

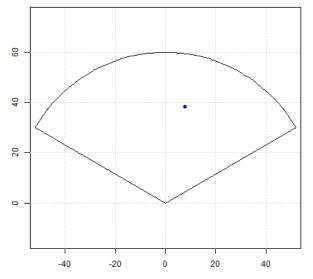
Mean absolute error: 2.16 m (95% CIs = 2.01 – 2.32)

Hastie et al (2019). Aquatic Conservation: Marine and Freshwater Ecosystems.

Classify marine mammals







X (m)

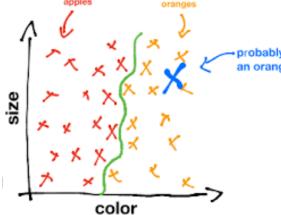
Classify marine mammals

Mobile targets in tidally energetic environments



- ➢ 95 targets.hr⁻¹
 - ✤ 6.6 seals.hr⁻¹
 - ✤ 88.8 non-seals.hr⁻¹
- Each day = ~2,100 non-seal targets;
- Need an effective means of data reduction.

kernel Support Vector Machines



The objective is to train a classification model based on labelled data. The trained model is then used for classifying novel data.

Kernel support vector machines

Used for a wide range of pattern recognition applications in biology



Quantifying movement behaviour of cheetahs from GPS tags;

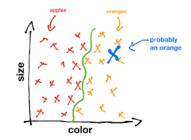
Constructing social networks based on co-occurrences of jackdaws;

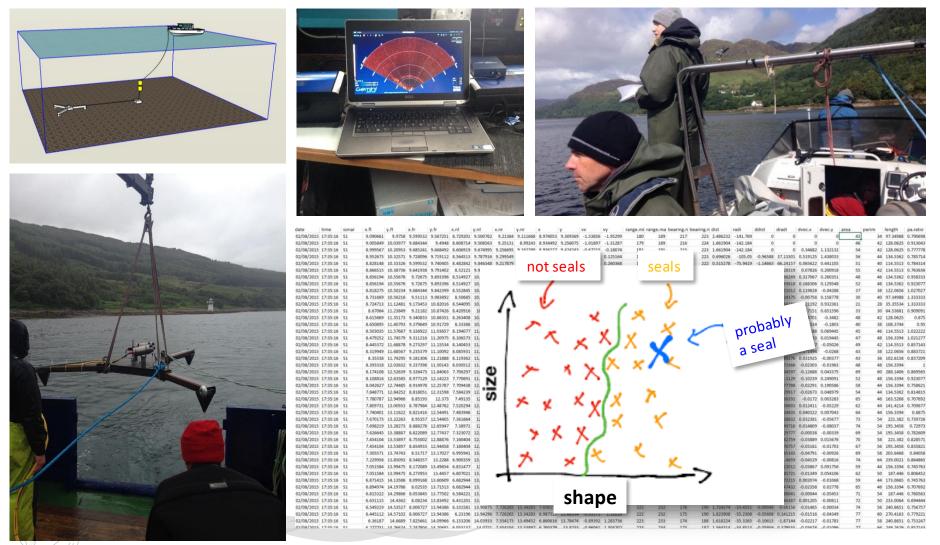
Counting individual wildebeest within aerial survey photo;

Classify seals in sonar data?



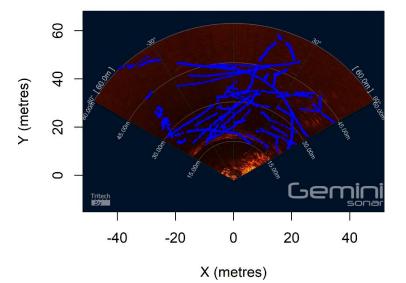
Kernel support vector machines



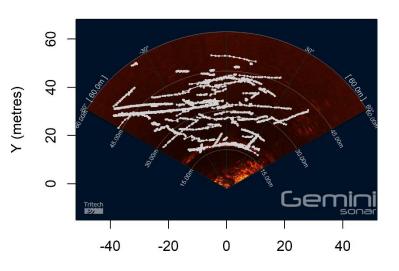


Kernel support vector machines





	Classified Seal	Classified Non-seal
Confirmed seals	100%	0%
Non-seal targets	8%	82%



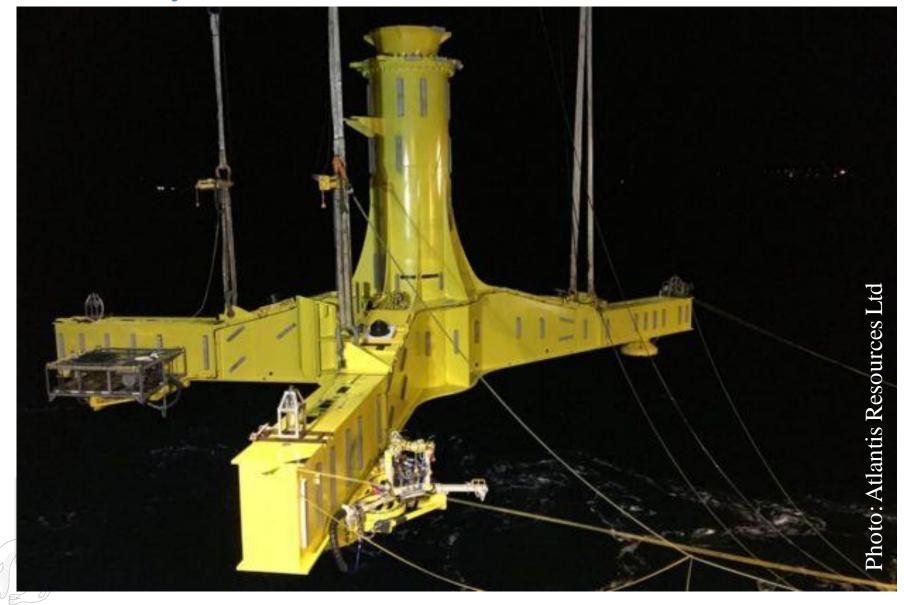
X (metres)

- All seals correctly classified;
- Majority of non-seals correctly classified;
- \geq ~8% of non-seals classified as seals.

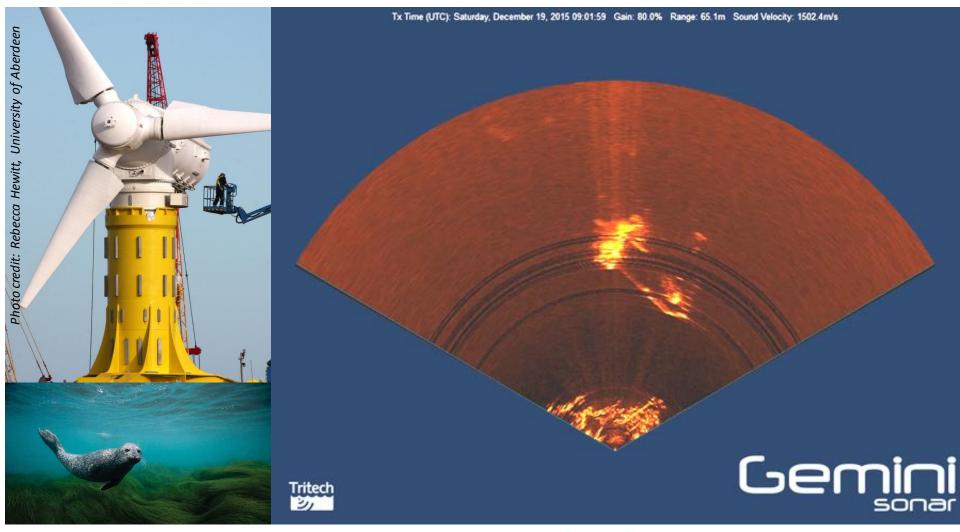
From: ~**89** false positive detections/hour To: ~**8** false positive detections/hour

Hastie et al (In press). Aquatic Conservation: Marine and Freshwater Ecosystems.

Seal tracking with sonar: Summary



Next steps: redeployment around operating turbine...











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